

Skyglow extends into the world's Key Biodiversity Areas

Joanne K. Garrett^{a,b*}, Paul F. Donald^{c,d}, Kevin J. Gaston^{a,e}

^a*Environment & Sustainability Institute, University of Exeter, Penryn, Cornwall TR10 9FE, U.K.*

^b*European Centre for Environment and Human Health, University of Exeter Medical School, Knowledge Spa, Royal Cornwall Hospital, Truro, Cornwall TR1 3HD, U.K.*

^c*Birdlife International, David Attenborough Building, Pembroke Street, Cambridge CB2 3QZ, U.K.*

^d*Conservation Science Group, Department of Zoology, University of Cambridge, Downing Street, Cambridge CB2 3EJ, U.K.*

^e*Wissenschaftskolleg zu Berlin, Institute for Advanced Study, Wallotstrasse 19, 14193, Berlin, Germany*

*corresponding author: Joanne K. Garrett, European Centre for Environment and Human Health, University of Exeter Medical School, Knowledge Spa, Royal Cornwall Hospital, Truro, Cornwall TR1 3HD, U.K.; J.K.Garrett@exeter.ac.uk

Short title: Skyglow extends into world's Key Biodiversity Areas

27 **Abstract**

28 The proportion of the Earth's surface that experiences a naturally dark environment at night is
29 rapidly declining with the introduction of artificial light. Biological impacts of this change have
30 been documented from genes to ecosystems, and for a wide diversity of environments and
31 organisms. The likely severity of these impacts depends heavily on the relationship between the
32 distribution of artificial nighttime lighting and biodiversity. Here, we carry out a global assessment
33 of the overlap between areas of conservation priority and the most recent atlas of artificial skyglow.
34 We show that of the world's Key Biodiversity Areas (KBAs), less than a third have completely
35 pristine nighttime skies, about a half lie entirely under artificially bright skies, and only about a fifth
36 contain no area in which nighttime skies are not polluted to the zenith. The extent of light pollution
37 of KBAs varies by region, affecting the greatest proportion of KBAs in Europe and the Middle
38 East. Statistical modelling revealed associations between light pollution within KBAs and
39 associated levels of both gross domestic product and human population density. This suggests that
40 these patterns will worsen with continued economic development and growth in the human
41 population.

42

43 Keywords: Artificial lighting, Atmosphere, Biodiversity, Nighttime, Streetlights

44

45 **Introduction**

46 The erosion of the nighttime through the introduction of artificial lighting, from street-lighting and
47 other sources, has pervasive environmental impacts. These span changes in the physiology and
48 behaviour of individual organisms, in the abundance and distribution of species, in the structure and
49 functioning of ecological communities, and in the provision of ecosystem services (Gaston *et al.*,
50 2013, 2014). A wide diversity of terrestrial, freshwater and marine organisms is influenced,
51 including microbes, plants and many groups of animals (e.g. crustaceans, molluscs, insects, fish,
52 amphibians, reptiles, birds, mammals; Gaston *et al.*, 2013; Bennie *et al.*, 2016).

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To date, attention has focussed foremost on the environmental impacts of the direct emissions from sources of nighttime lighting. This is, however, only relatively narrowly spatially distributed compared with the skyglow that is caused by upwardly emitted or reflected artificial light being scattered in the atmosphere by water, dust and gas molecules. The latter has been estimated already to extend over ~23% of the global land area (Falchi *et al.*, 2016). It can increase background sky brightness to levels comparable to those of late twilight and moonlight, and can obscure the visibility to humans of individual stars and the Milky Way (83% of the human population lives under light polluted skies; Falchi *et al.*, 2016). It is likely to interfere with multiple biological processes including activity patterns of diurnal, crepuscular and nocturnal species (which are variously sensitive to timing of twilight and nighttime levels of moonlight; e.g. Moore *et al.*, 2000; Bachleitner *et al.*, 2007), and nighttime orientation and navigation (which often involves the use of stars and other celestial objects; e.g. Foster *et al.*, 2017). Some free-living organisms are able to detect and respond to extraordinarily low levels of nighttime light (e.g., Warrant *et al.*, 2004).

Particularly because of the contribution of light that is emitted, or reflected, at relatively shallow angles to the horizontal, skyglow can extend substantial distances (up to hundreds of kilometers) from urban sources of nighttime lighting (Luginbuhl *et al.*, 2014). This raises the potential for impacts of artificial nighttime lighting to reach many globally important biodiversity areas, even when these are reasonably remote from many other anthropogenic pressures. In this paper, we estimate the extent of this overlap, based on the most recent global modelling of skyglow and the distribution of Key Biodiversity Areas (KBAs).

77 **Materials and methods**

78 **Data**

79 Estimates of global variation in skyglow were obtained from Falchi *et al.* (2016). This surface was
80 produced by the modelling of measured upward radiance from artificial sources from satellite
81 imagery (from the VIIRS DNB sensor on the Suomi National Polar-orbiting Partnership (NPP)
82 satellite), and ground measurements. The data are presented for the entire area (terrestrial and
83 marine) between approximately 85°N and 60°S, at a spatial resolution of 30-arcseconds (~1 km) as
84 a ratio of artificial brightness to natural brightness (typical night sky background excluding the
85 brightest stars and the Milky Way). The authors define the level of artificial brightness under which
86 a sky can be considered “pristine” as up to 1% above the natural background level (ratio of 0.01).
87 At a level of 8% or more above natural conditions (ratio of 0.08) light pollution extends from the
88 horizon to the zenith and the entire sky can be considered polluted. We use these two thresholds in
89 our analysis.

90
91 Key Biodiversity Areas (KBAs) are sites that contribute significantly to the global persistence of
92 biodiversity, and use quantitative criteria to identify places that support viable populations of
93 species for which site-scale conservation is appropriate (IUCN, 2016). The current inventory
94 includes Important Bird and Biodiversity Areas (IBAs), KBAs identified during Biodiversity
95 Hotspot profiles supported by the Critical Ecosystem Partnership Fund (CEPF) and Alliance for
96 Zero Extinction sites (AZEs) (<http://www.keybiodiversityareas.org/home>). We used boundary data
97 (BirdLife International, 2016) for all 14,979 KBAs for which polygons were available in December
98 2016 (comprising >95% of all KBAs recognised in March 2018). Only those that lie within the
99 extent of the skyglow data were considered further (14,765). These had a combined coverage of
100 34,785,579 sq km, with a median area of 209 sq km (calculated in ArcGIS 10.3.1, ESRI Inc.;
101 WGS84 datum). The KBA dataset includes characterisation by region, and this includes a “Marine”
102 category identifying those which are predominantly marine based.

103

104 Data were also obtained on three further variables that are of potential significance as determinants
105 of the levels of skyglow experienced by KBAs (excluding those on the ‘High seas’): Gross
106 Domestic Product (GDP; which provides a measure of economic activity), human population
107 density, and protected area coverage: (i) Median GDP per capita, adjusted for purchasing power
108 parity (PPP), was extracted from the gridded (spatial resolution of 5 arc-min) product by Kummu *et*
109 *al.* (2018). GDP grid cells were included in the median calculation if the centre was within the KBA
110 boundary (missing $n = 792$; 95 % of KBAs retained for further analysis). This product uses sub-
111 national datasets where possible in combination with national datasets (including the World Bank
112 Development Indicators database and the CIA World Factbook for missing data). (ii) Median
113 human population density for each KBA was extracted from the gridded population density data
114 (resolution of ~ 1 km) of the world projected for the year 2015 from census data (CIESIN, 2016)
115 where data were included when the centre of the grid cell was located inside the KBA boundary
116 (missing $n = 291$). And, (iii) protected area coverage for each KBA was calculated using the World
117 Database of Protected Areas (IUCN and UNEP-WCMC, 2016). Protected areas were only included
118 where they had associated spatial boundary data.

119

120 **Data analysis**

121 Analyses were carried out in the software package R (version 3.1.3) (R Development Core Team,
122 2014) using the packages “raster” (version 2.5.2) and “rgdal” (version 1.1.8). Ratio values of
123 artificial sky brightness were extracted from the global dataset for each KBA. Sky brightness values
124 were stored as a grid (raster) with a spatial resolution of 30-arcseconds (~ 1 km). Values were
125 included when the centre of the corresponding cell was located within the KBA boundary. Two
126 measures were calculated for each KBA: (i) percentage coverage by pristine nighttime skies (ratio
127 values of ≤ 0.01) and (ii) percentage coverage by skies not polluted from the horizon to zenith (ratio

128 values <0.08). In addition, we calculated the percentage coverage of the total area of KBAs by these
129 respective levels of sky brightness.

130

131 The percentiles (0 - 100) were calculated for both the median KBA GDP (per capita, PPP, thousand
132 \$) and median KBA population density (people per sq km). The median proportion pristine for the
133 KBAs with each combination of GDP and population density percentile was then calculated and
134 plotted. The use of percentiles maximises the evenness of the sample sizes for each combination.

135

136 We used a generalised linear model (GLM), with a binomial error structure and logit link function
137 to model whether a KBA had entirely pristine skies or not as a function of GDP per capita,
138 population density, the interaction between the two, and the proportion of the KBA that falls within
139 a protected area. To account for strong right skew in population density, population density was
140 categorised on a log scale (0 - 1 people per sq km, $>1 - 10$, $>10 - 100$, $>100 - 1,000$ and $>1,000$).
141 Proportion protected had a bimodal distribution with high proportions of KBAs either fully or not
142 protected at all. This variable was dichotomised as either fully protected (100 % protected) or not
143 (<100 % protected). Model fit is given by McFadden's *pseudo- R^2* . Although there was significant
144 pair-wise correlation between the predictor variables, the generalised variance inflation factors,
145 standardised by degrees of freedom, were all <2 (Supp. Tables 1-2).

146

147 **Results**

148 A half (51.5%) of the total number of KBAs assessed contained no area with pristine nighttime
149 skies while less than one third (29.5%) had completely pristine nighttime skies (ratio values $\leq 0.01 \approx$
150 up to 1 % above natural conditions; Fig. 1a). Europe had the greatest percentage (81%) of KBAs
151 containing no area of pristine skies, followed by the Middle East (75%) and the Caribbean (64%;
152 Fig. 2a). The only regions in which all KBAs had completely pristine skies were Antarctica and
153 Marine (Fig. 2a).

154

155 About one fifth (21.0%) of all KBAs consisted entirely of area in which night skies were polluted to
156 the zenith (Fig. 1b). However, over a half of KBAs (51.9%) are completely free of skies polluted to
157 the zenith (ratio values $<0.08 \approx$ up to 8 % above natural conditions; Fig. 1b). The Middle East was
158 the region where the greatest percentage (46 %) of KBAs had nighttime skies entirely polluted to
159 the zenith, followed by Europe (34%) and the Caribbean (32%; Fig. 2).

160

161 Of the summed global area of KBAs, more than 15% was not pristine and more than 5% was
162 polluted to the zenith (Fig. 2b). The Middle East was the region with the greatest percentage (54%)
163 of the overall KBA area not having pristine skies, followed by Europe (53%) and the Caribbean
164 (33%; Fig. 2b). These were also regions with the largest percentage of the overall KBA area that is
165 polluted to the zenith (Middle East - 25%, Europe - 19%, Caribbean - 10%; Fig. 2b).

166

167 The likelihood of a KBA having pristine skies decreased with increasing GDP and population
168 density, and the interaction between the two, and increased with proportional coverage by protected
169 areas (Table 1). Neither the categories 1 - 10 people per sq km or 10 - 100 people per sq km were
170 found to be significantly different to 0 - <1 people per sq km. Where there were >100 to 1,000
171 people per sq km, KBAs were less than half as likely to be pristine (compared to 0 - <1 people per
172 sq km; Odds ratio (OR) = 0.44, 95 % Confidence interval (CI) = 0.34 - 0.56). Where there were
173 $>1,000$ people per sq km, the likelihood that the KBA would be entirely pristine was extremely low
174 (OR = 0.02, 95 % CI = 0.01 – 0.06). With every increase in GDP by \$1,000 there was an associated
175 decrease in likelihood of being pristine by ~1 % (OR = 0.99, 95 % CI 0.98 - 0.99). KBAs which
176 were fully protected were 1.2 x as likely to be pristine compared to those not fully protected (OR =
177 1.24, 95 % CI = 1.05 - 1.7). There was also a significant interaction between population density and
178 GDP (Table 1; Fig. 3).

179

180 **Discussion**

181
182 Skyglow is often envisaged as an exclusively urban issue. However, both modelling and ground
183 measurements have shown it to be very widespread, often being propagated over long distances
184 from sources (Kyba *et al.*, 2015; Falchi *et al.*, 2016). Nevertheless, it is perhaps surprising that less
185 than a third of KBAs had completely pristine nighttime skies, more than a half contained no area
186 with pristine skies, and that a sixth of the total area of KBAs was light polluted. This is especially
187 so considering that the global data on skyglow are likely, if anything, to be conservative estimates
188 of its extent. Such data are primarily estimated using satellite measurements and, as such, are for
189 ‘open sky’ conditions (cloud-free). They therefore do not take into account the amplification of
190 skyglow that can occur by cloud cover (Jechow *et al.*, 2017), and likely under-represent the
191 occurrence of artificial brightness of horizons (which may be important for many organisms, for
192 example in influencing predator-prey interactions).

193
194 These results are especially troubling because it has become increasingly apparent that organisms
195 can respond even to absolutely small (which may nonetheless be relatively large) changes in natural
196 nighttime light conditions (Warrant *et al.*, 2004), in perceived day lengths (Gaston *et al.*, 2017), and
197 in artificial nighttime lighting (Gaston *et al.*, 2014). The breadth and number of species whose
198 behaviour is influenced by skyglow seems likely to be large.

199
200 Unsurprisingly, the likelihood of the skies of a KBA experiencing skyglow tends to increase in
201 countries with higher GDP, and in areas with higher human population density, consistent with
202 global results (Gallaway, Olsen & Mitchell, 2010). It is possible to have KBAs in regions with
203 relatively pristine skies in areas with high human densities when these populations are economically
204 poor, and likewise in areas with high GDP when these populations are at low density (Fig. 4).
205 However, under virtually all other circumstances KBAs exist under light polluted skies. This

206 strongly suggests that globally the proportion of KBAs experiencing skyglow will almost certainly
207 increase in parallel with developing country economies.

208

209 The relatively small areal extent of many KBAs means that most commonly their skies are either
210 entirely light polluted or entirely unpolluted. Whether they have coverage by protected areas is also
211 associated with whether they experience skyglow, consistent with results for upwardly emitted light
212 (Gaston, Duffy & Bennie, 2015). However, this seems in most cases unlikely to be a consequence
213 of protection *per se*. More likely, it is a result of the tendency for protected areas to be distributed
214 away from urban centres (and hence sources of artificial light), and often in regions with reduced
215 competition over land use (Gaston *et al.*, 2008).

216

217 Here we have identified the regions where KBAs are most affected by skyglow. This effectively
218 prioritises areas for further detailed assessment of the potential risk to species of conservation
219 priority inhabiting these. To assess the risk, the extent of overlap between skyglow and species
220 ranges and occupancy could be calculated. A similar technique has been applied to cacti and
221 mammal species ranges, where the overlap with light pollution was calculated for upwardly emitted
222 light radiation measured directly from satellites (Duffy *et al.*, 2015; Correa-Cano *et al.*, 2018).
223 Further research could also identify areas currently minimally affected by light pollution but at risk
224 of increases in future due to rapidly developing economies or with increasing populations,
225 facilitating the development of mitigation measures in these areas.

226

227 Dramatic reductions in anthropogenic pressures on biodiversity are often costly to achieve, and
228 there are commonly substantial lag times between such reductions and biodiversity responses. By
229 contrast, marked reductions in skyglow could be achieved by limiting outdoor artificial lighting to
230 levels and places where it is required by people, which would result in considerable cost savings

231 without undermining the benefits that it brings. Indeed, one might argue that environmental benefits
232 and financial savings from considered lighting policies are closely aligned (Gaston, 2013).

233

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237

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 298
 299

300 **Table 1**

301 Results of GLM (binomial) of relationship between whether a Key Biodiversity Area had entirely
 302 pristine skies or not and per capita GDP (median within KBA), human population density, the
 303 interaction between the two, and the proportion of the KBA that falls within a protected area.

304

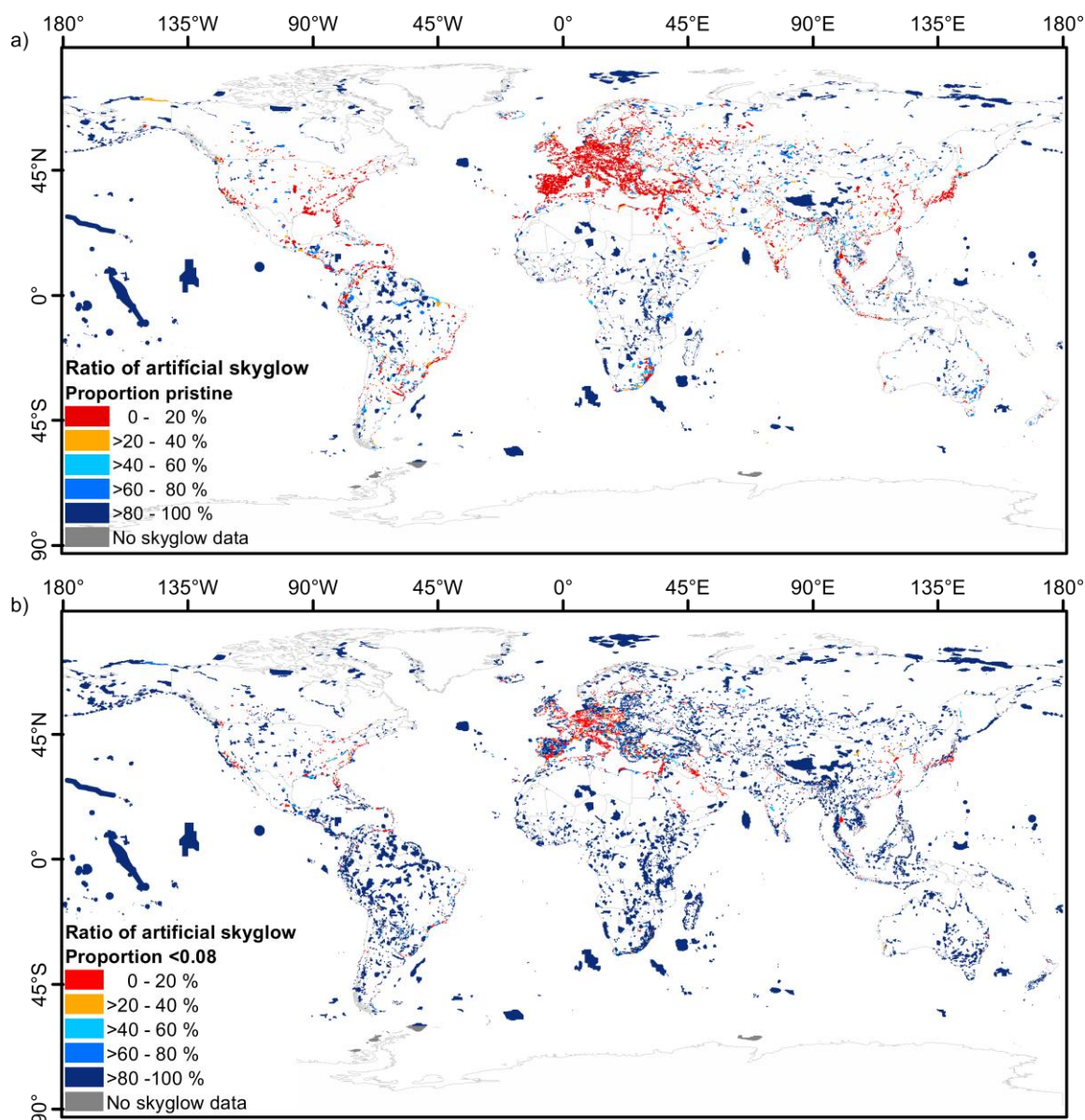
Variable	Estimate	Standard error	Z value	<i>p</i>	OR	95 % CI
Population density (people per sq km)						
<i>> 1,000</i>	-3.82	0.52	-7.42	<0.001	0.02	0.01 - 0.06
<i>>100 - 1,000</i>	-0.83	0.13	-6.37	<0.001	0.44	0.34 - 0.56
<i>>10 - 100</i>	0.15	0.11	1.40	0.161	1.16	0.94 - 1.44
<i>>1 - 10</i>	0.22	0.12	1.88	0.060	1.25	0.99 - 1.58
<i>0 - 1 (ref)</i>						
Median GDP (thousands)	-0.01	0.00	-3.73	<0.001	0.99	0.98 - 0.99
Proportion protected						
<i>100 %</i>	0.22	0.09	2.51	0.012	1.24	1.05 - 1.47
<i><100 % (ref)</i>						
Interaction population density x GDP						
<i>> 1,000 * GDP</i>	-0.04	0.04	-0.95	0.341	0.96	0.88 - 1.04
<i>>100 - 1000 * GDP</i>	-0.33	0.03	-10.83	<0.001	0.72	0.68 - 0.76
<i>>10 - 100 * GDP</i>	-0.24	0.01	-24.92	<0.001	0.79	0.77 - 0.80
<i>>1 - 10 * GDP</i>	-0.09	0.01	-13.82	<0.001	0.92	0.91 - 0.93
<i>0 - 1 (ref) * GDP</i>						
Intercept	0.47					
N	13913					
McFadden's <i>pseudo-R</i> ²	0.27					

305 OR = odds ratio

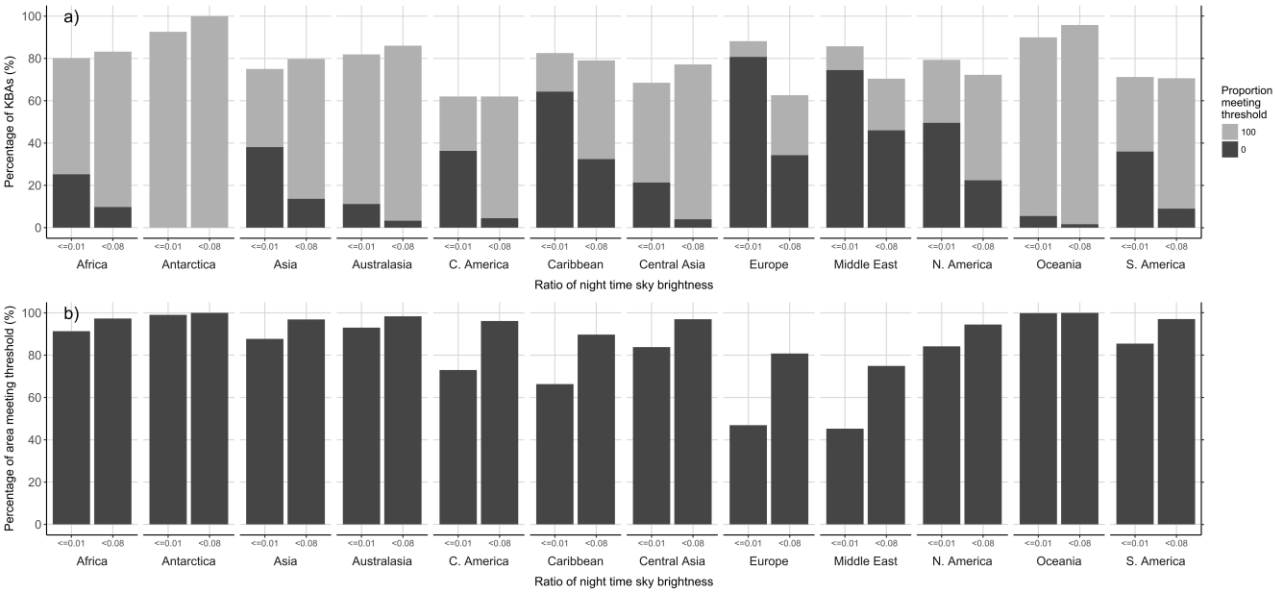
306 95 % CI = 95 % Confidence Interval

307

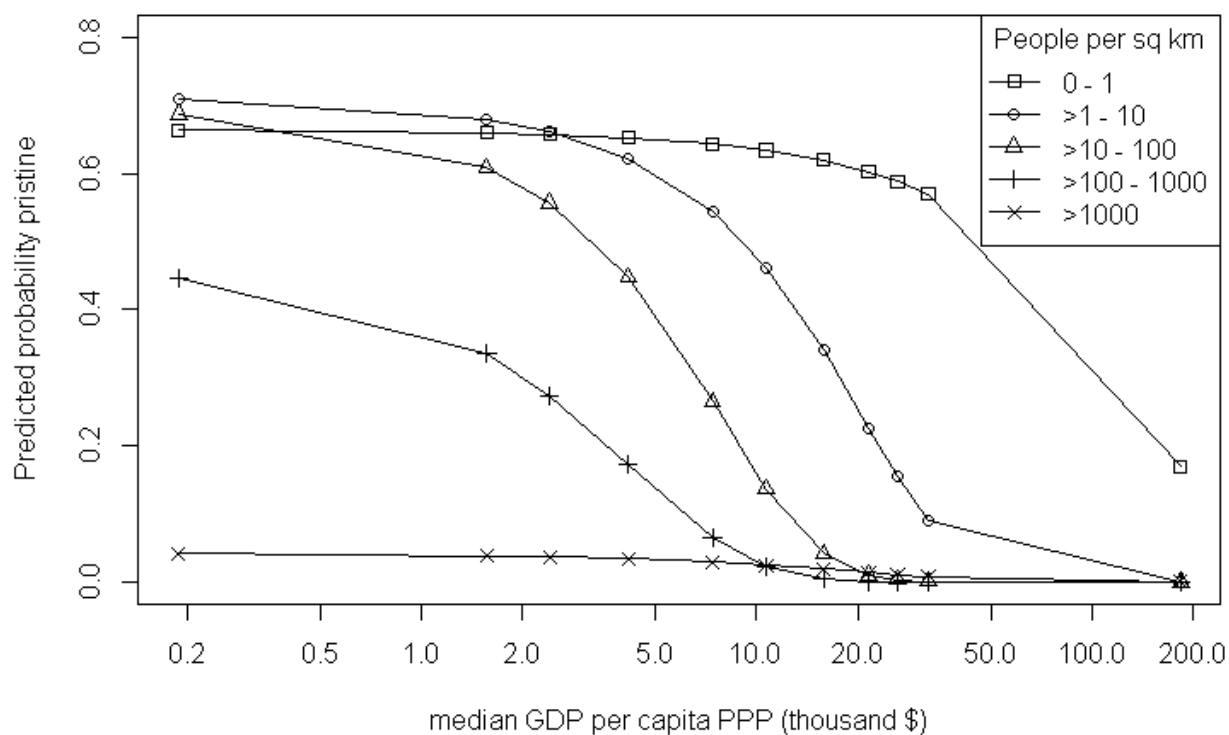
308 **Fig. 1.** The proportion of the extent of Key Biodiversity Areas with a) pristine nighttime skies (ratio
 309 of artificial brightness to natural brightness ≤ 0.01) and b) nighttime skies not polluted to the zenith
 310 (ratio of artificial brightness to natural brightness < 0.08). The outlines have been exaggerated for
 311 display purposes.



315 **Fig. 2.** a) Proportion of Key Biodiversity Areas which have 0% and 100% coverage of pristine
 316 nighttime skies (ratio threshold of artificial brightness to natural brightness ≤ 0.01) and skies not
 317 polluted to the zenith (ratio threshold of artificial brightness to natural brightness < 0.08). b) Total
 318 proportion of area with pristine night time skies and skies not polluted to the zenith (ratio of
 319 artificial brightness to natural brightness < 0.08) (bottom). Skyglow ratio thresholds are specified on
 320 the x axis. Regions displayed in alphabetical order, the marine region is excluded as it is 100 %
 321 pristine.



332 **Fig. 3.** Predicted probabilities of KBA being pristine from GLM model results as a function of GDP
 333 and population density, and fixed proportion protected of 100 %. Note the interaction between
 334 population density of >1,000 and GDP was not significant.



348 **Fig. 4.** Median proportion of Key Biodiversity Areas that has pristine nighttime skies for each
349 combination of GDP and population percentile (0-100).

